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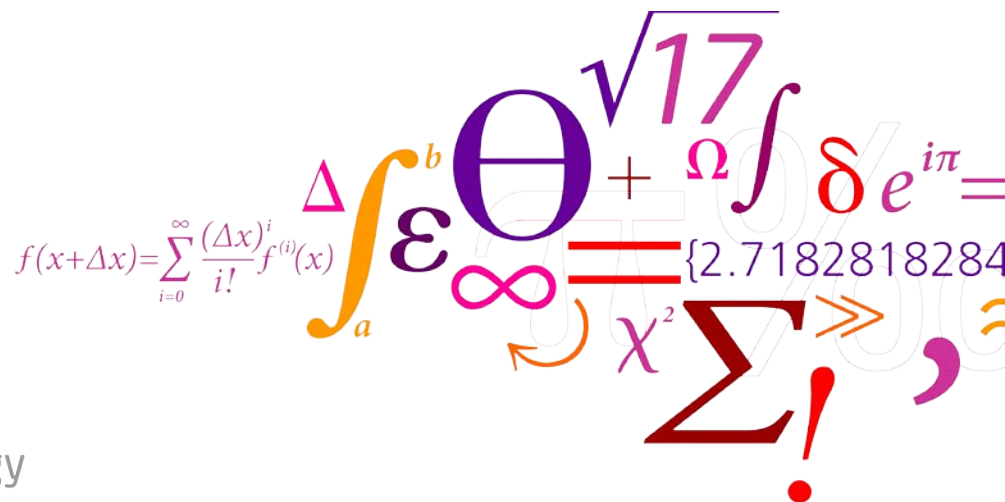
From trades to turbines – linking global, mesoscale, and local models

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Jake Badger, Alfredo Peña, Xiaoli Larsen, Claire Vincent,
Caroline Draxl, Mark Kelly, Jens Carsten Hansen and Niels
Gylling Mortensen



$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\int_a^b \varepsilon \Theta^{\sqrt{17}} + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

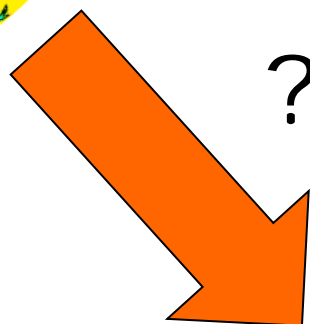
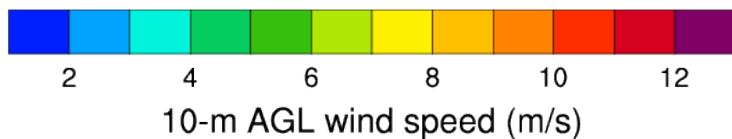
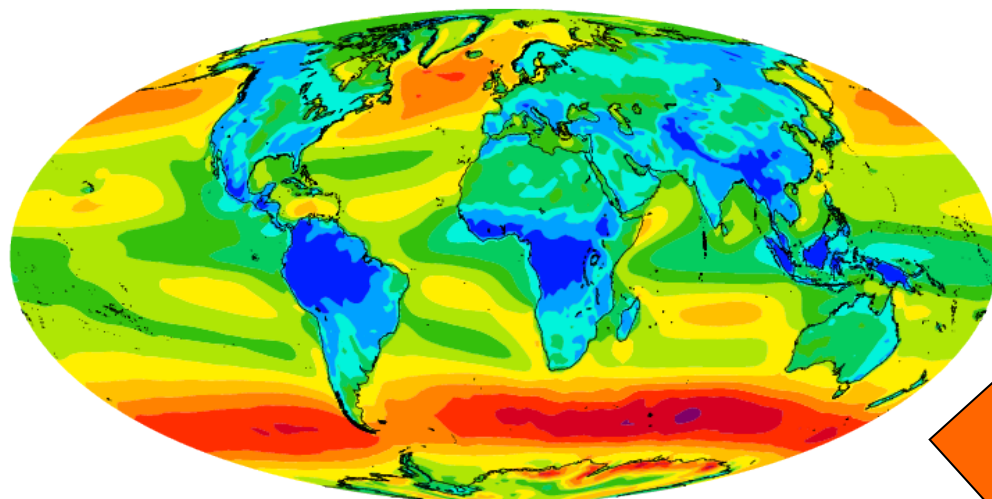
$$\chi^2 \sum ! >>$$

Risø DTU

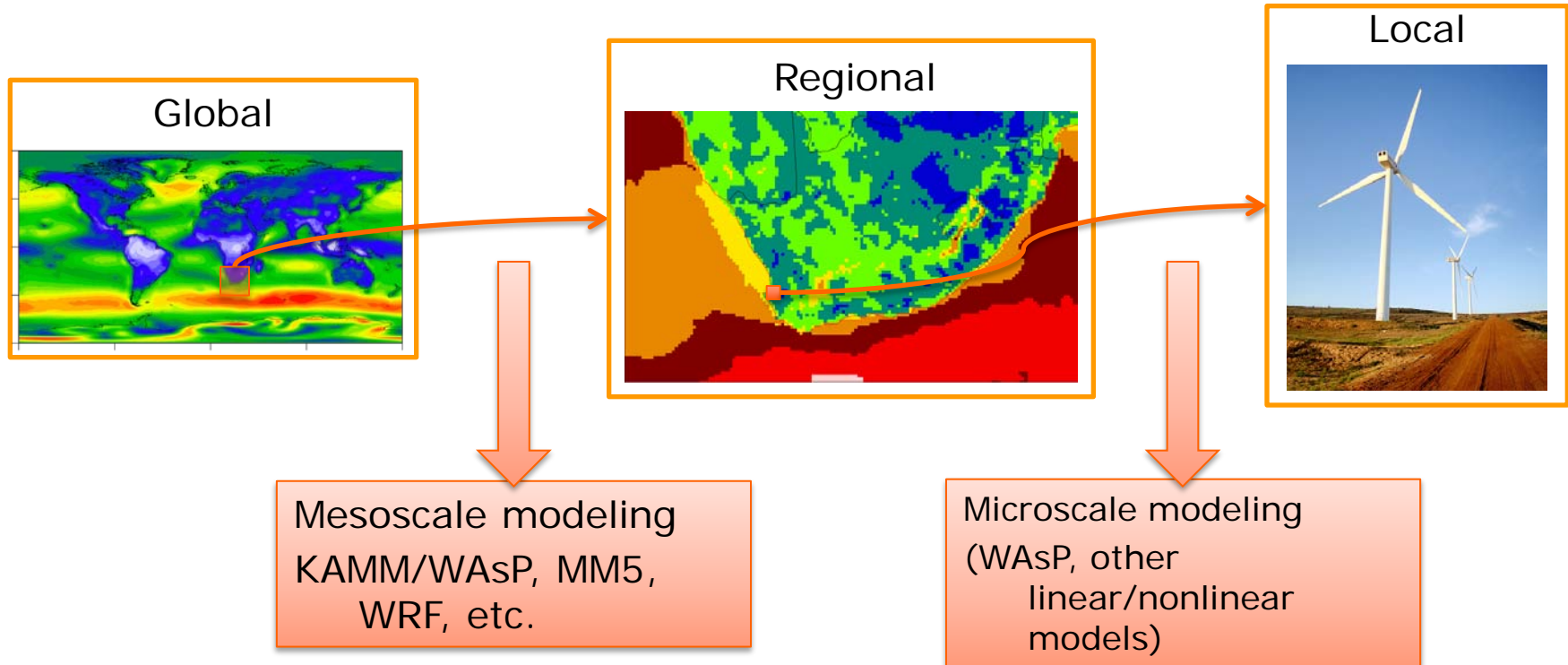
National Laboratory for Sustainable Energy

Outline

- The problem – an introduction
- From large-scale to mesoscale
 - Statistical-dynamical downscaling
 - Dynamical downscaling
- From mesoscale to microscale
 - The effects of resolution
 - How to use the mesoscale model information
 - Generalization
- Implications for verification
- Other applications
- Summary



Numerical Wind Atlas - Downscaling steps

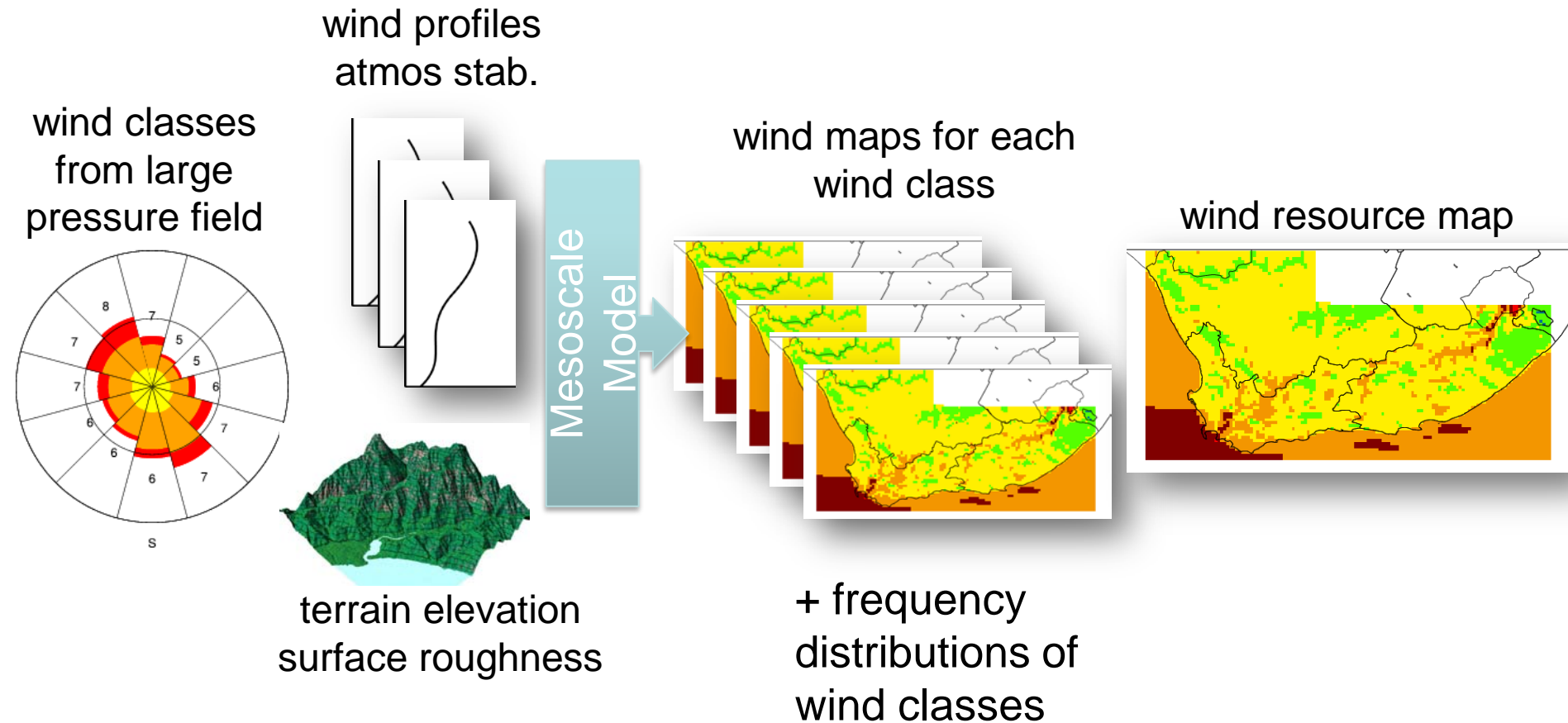


For now we assume that the models are perfect, and concentrate on their coupling

KAMM: Karlsruher non-hydrostatic mesoscale model

WAsP: Wind Atlas Analysis and Application (widely used wind resource tool)

From large-scale to mesoscale: statistical downscaling



Simple/Fast/Cheap

Complex/Slow/Expensive

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National Laboratory for Sustainable Energy

~~Interpolation~~

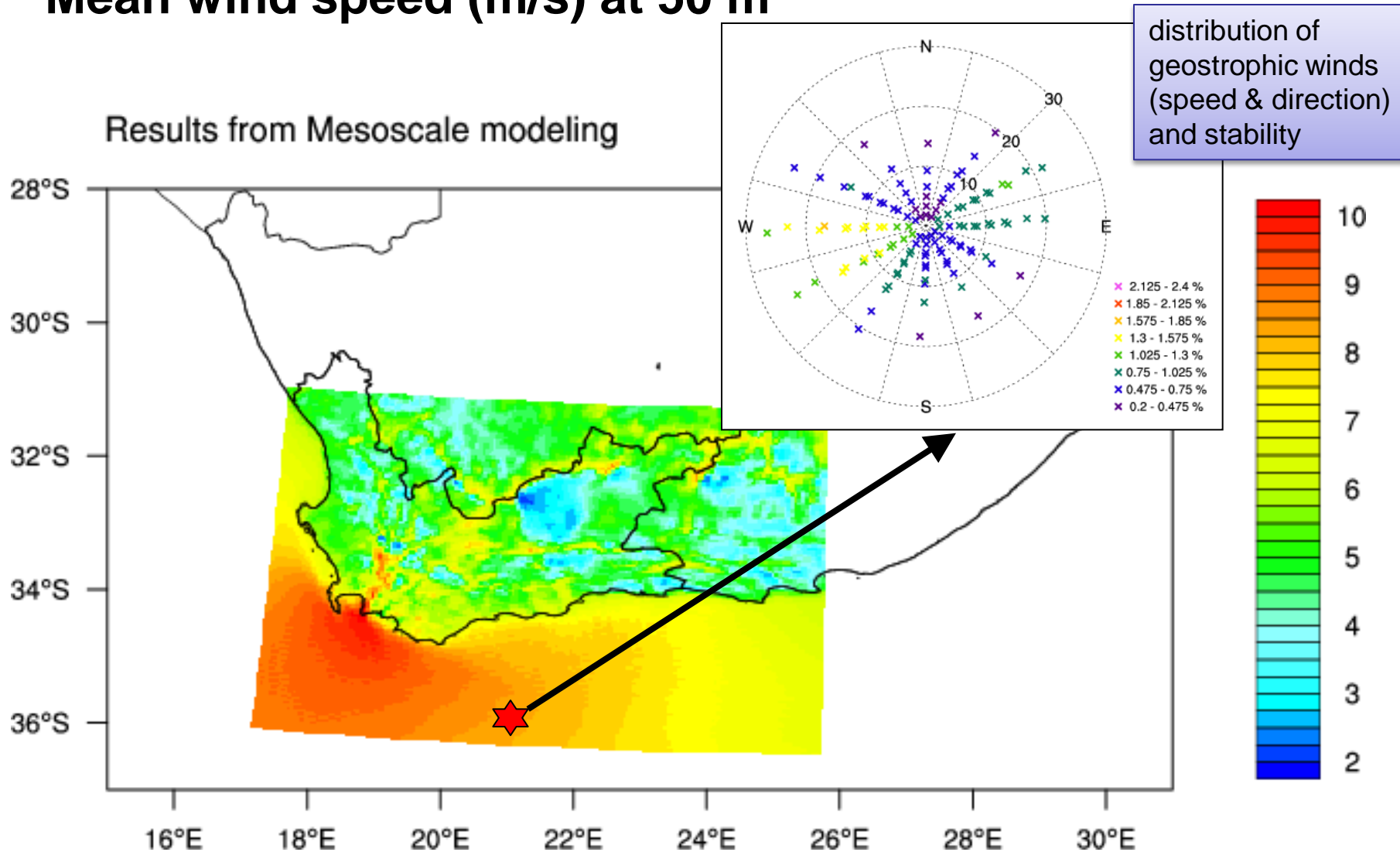
Risø Wind
Atlas

Statistical-
dynamical

Fully
dynamical

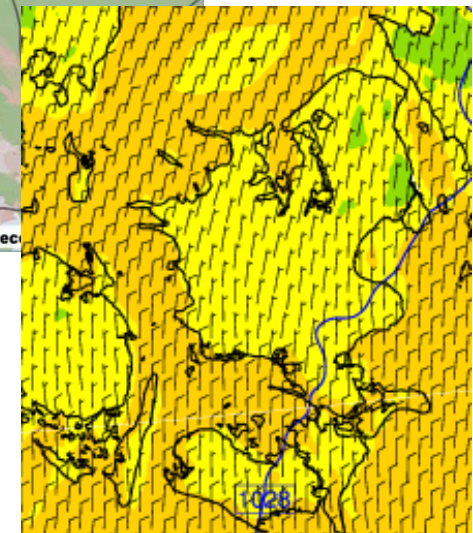
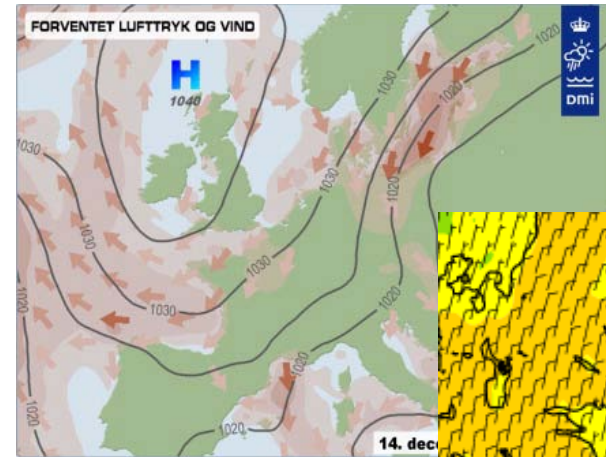
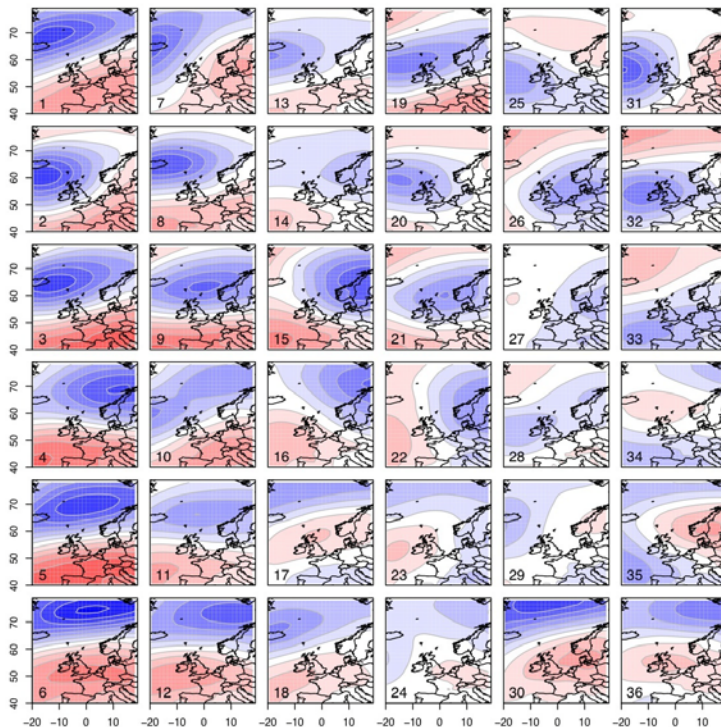
Preliminary calculations for South Africa

Mean wind speed (m/s) at 50 m



Assumptions used in statistical downscaling

- Regional wind climate can be adequately represented by the combination of a finite number of weather “states”
- There is a one-to-one relationship between each of these states and the local wind conditions

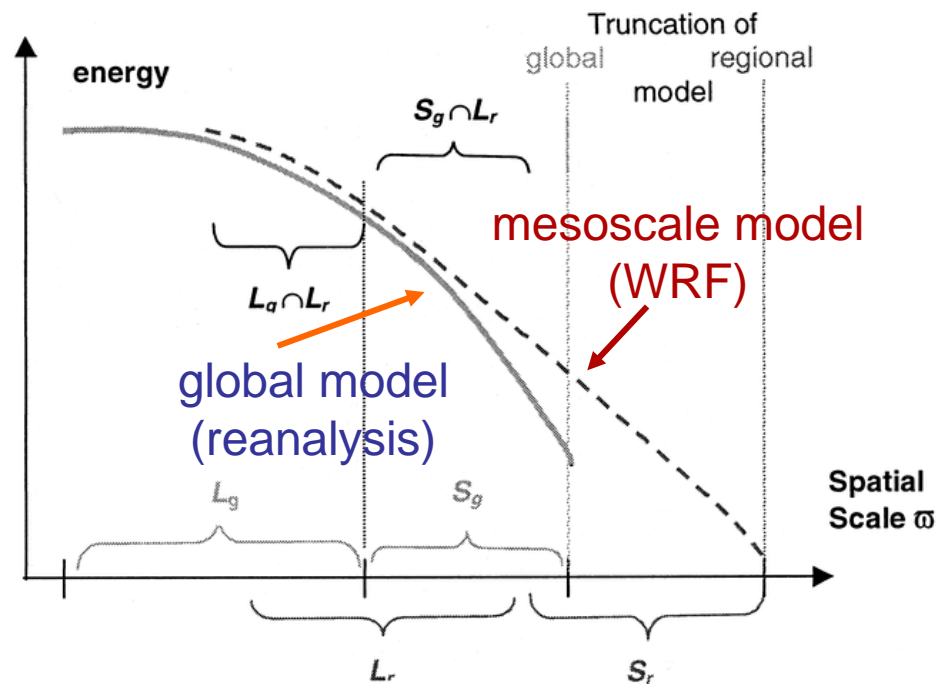


These assumptions break down in regions where strong (thermal) mesoscale forcing exists (sea-breeze, mountain drainage flow, etc.)

Dynamical Downscaling

- Not weather forecasting
- Not regional climate modeling

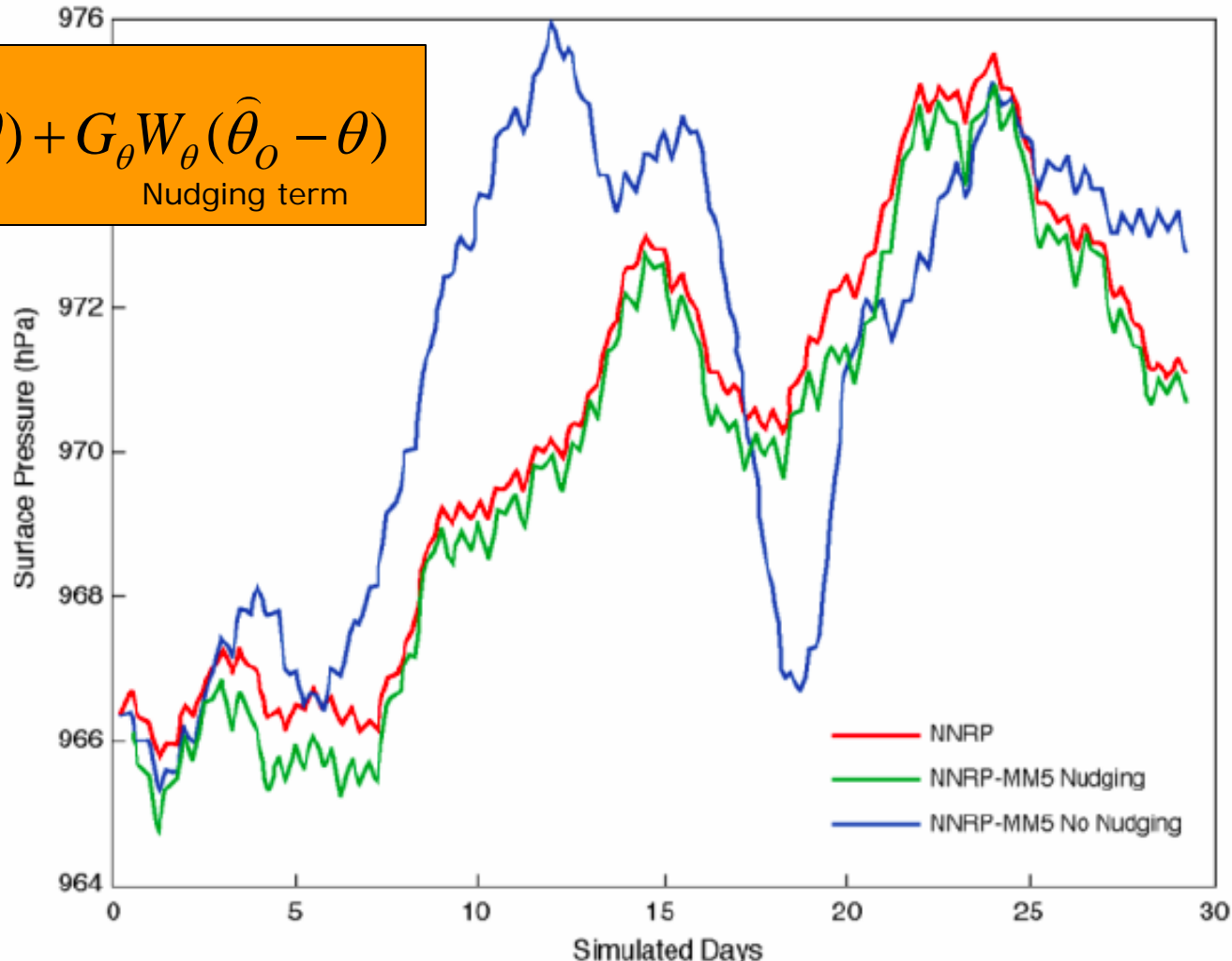
We “trust” the large-scale reanalysis from which the downscaling is based
 We need to resolve smaller scales not present in the reanalysis



Is the downscaling simulation in sync with the driving analysis?

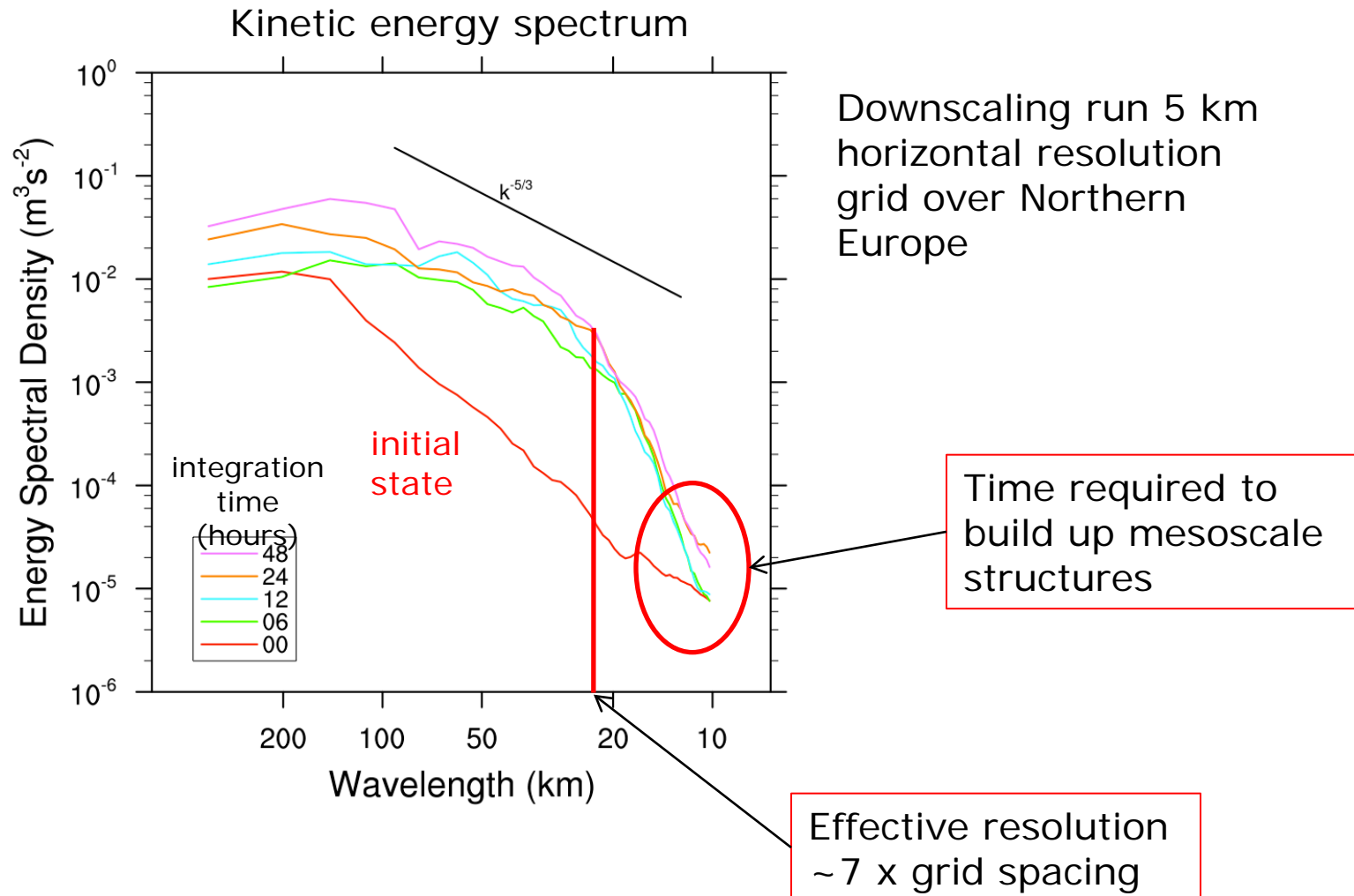
$$\frac{\partial \theta}{\partial t} = F(\theta) + G_{\theta} W_{\theta} (\hat{\theta}_o - \theta)$$

Nudging term



Domain-averaged surface pressure for a MM5 run over the Pacific Northwest (USA) - from Clifford Mass, Univ. of Washington

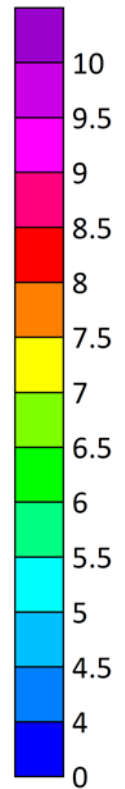
Spin-up and resolution effects



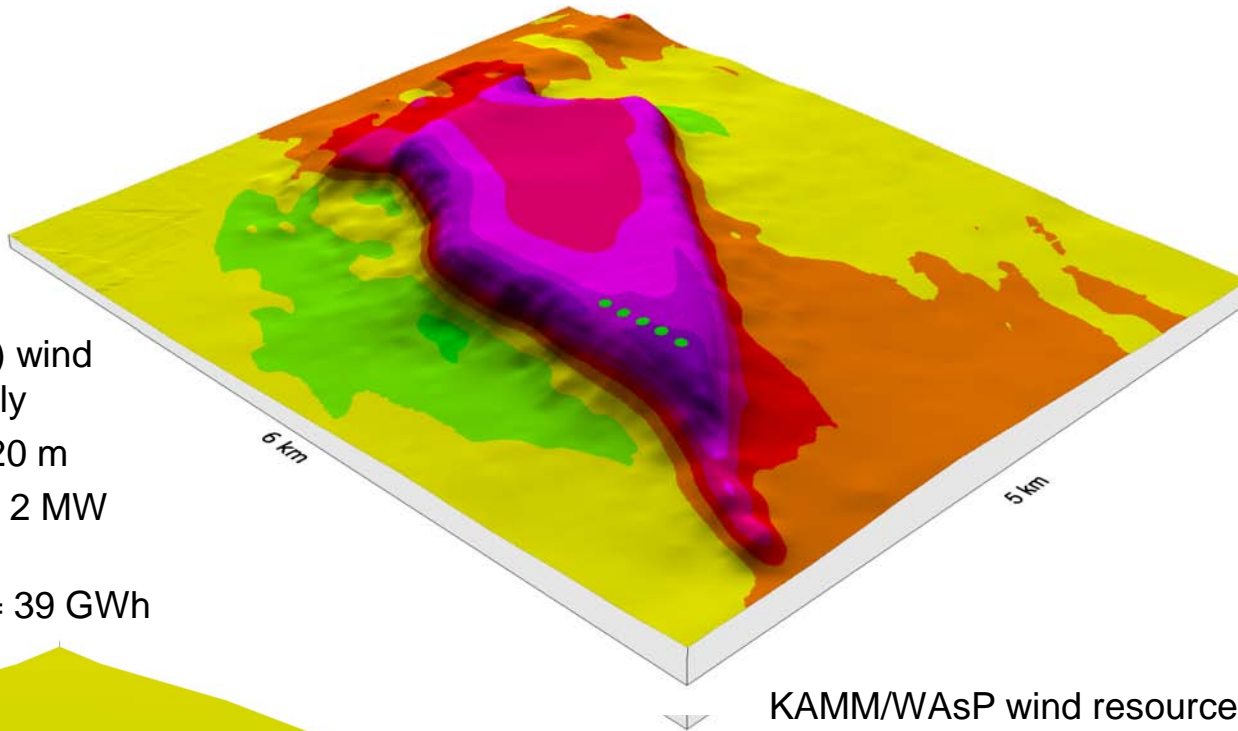
Many remaining issues...

- While dynamical downscaling is often preferred, many issues remain unresolved
 - nudging (strength, level, fields?) versus re-initialization (how often, spin-period length?)
 - length (or sampling strategy) of the simulations – do they capture the interannual (interdecadal) variability?
 - what is the adequate spatial resolution – small enough to capture detailed mesoscale structures, large enough for parameterizations to remain valid
 - since coupling to microscale – avoid double representing small-scale structures
 - ??

Need for mesoscale to microscale downscaling: Resolution is key in applications



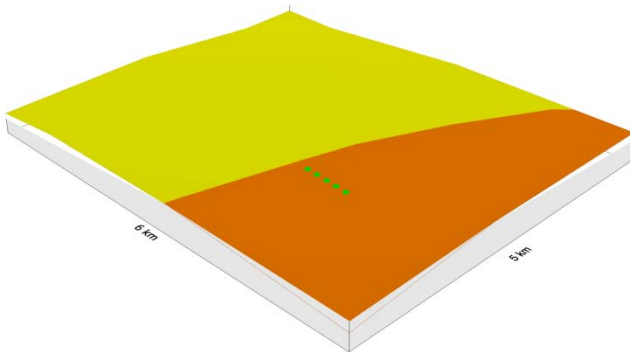
wind
speed
(m/s)



KAMM/WAsP wind resource map

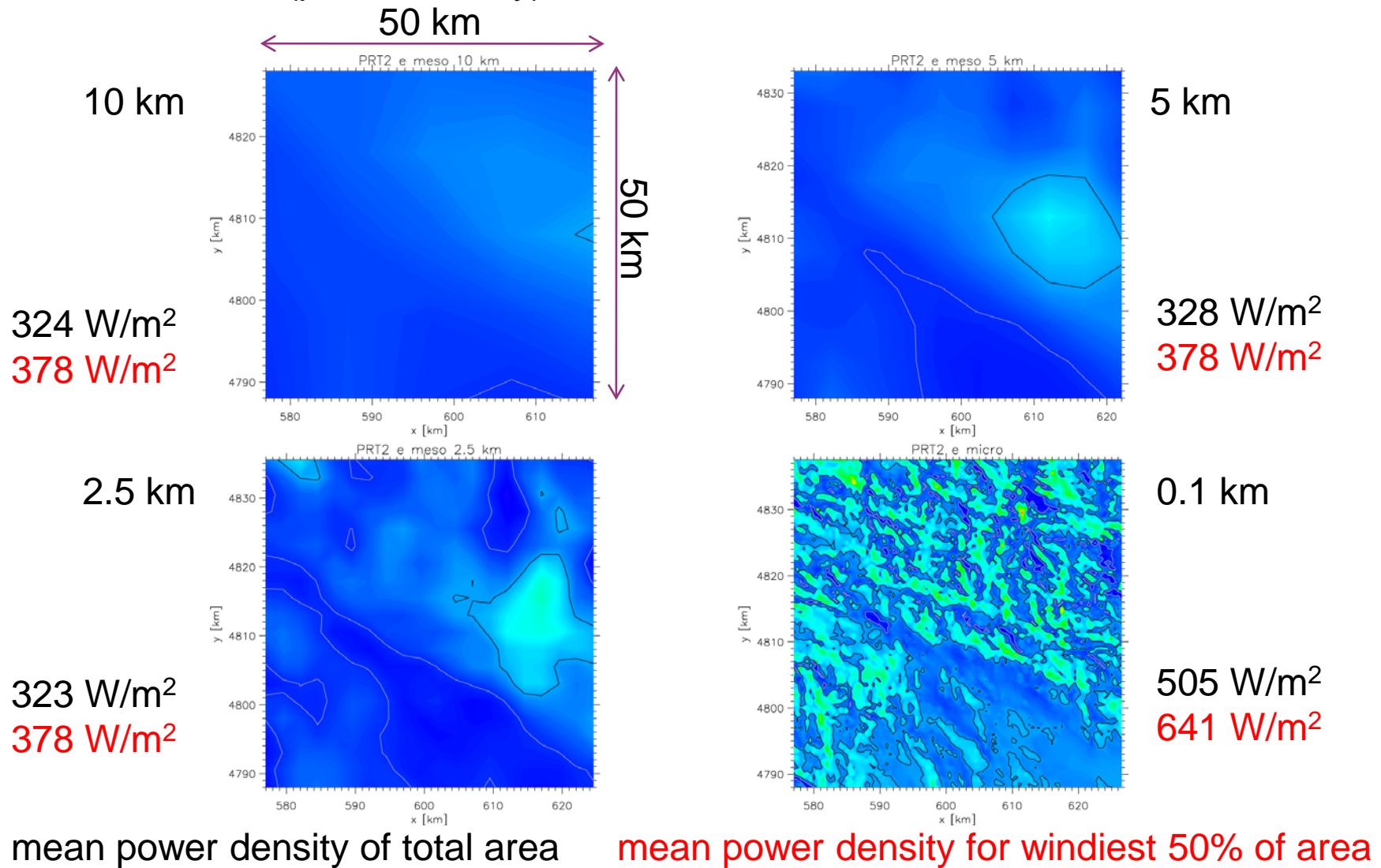
- Grid cell size 20 m
- Wind farm of five 2 MW turbines
- Estimated AEP = 55 GWh

- Mesoscale (KAMM) wind resource map only
- Grid cell size 5120 m
 - Wind farm of five 2 MW turbines
 - Estimated AEP = 39 GWh

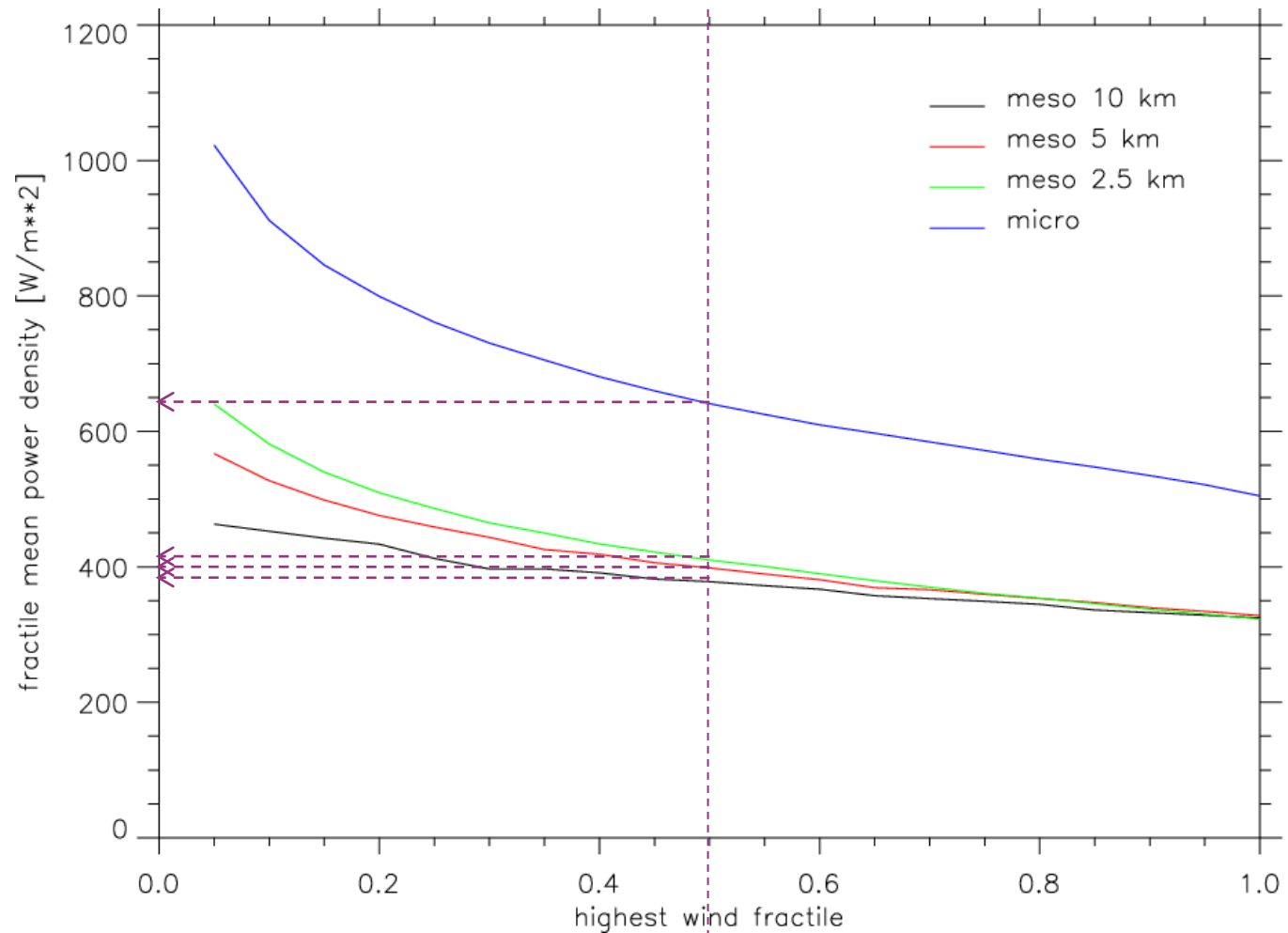


Importance of resolution

Wind resource (power density) calculated at different resolutions



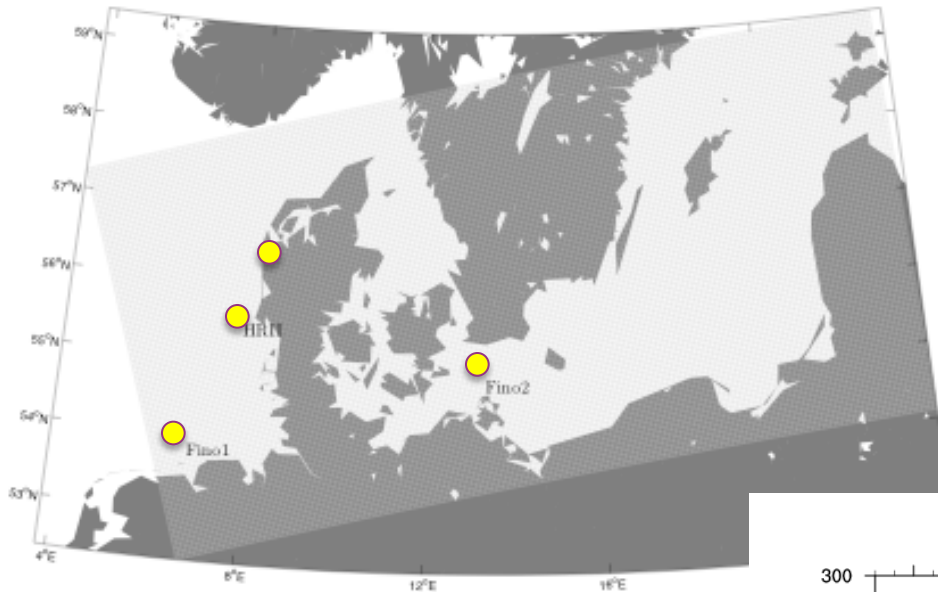
Importance of resolution



Note: this area exhibits very large topography effects. Even for Danish landscape effect can give 25 % boost in wind resource at the windiest 5 percentile.

Median power density windiest

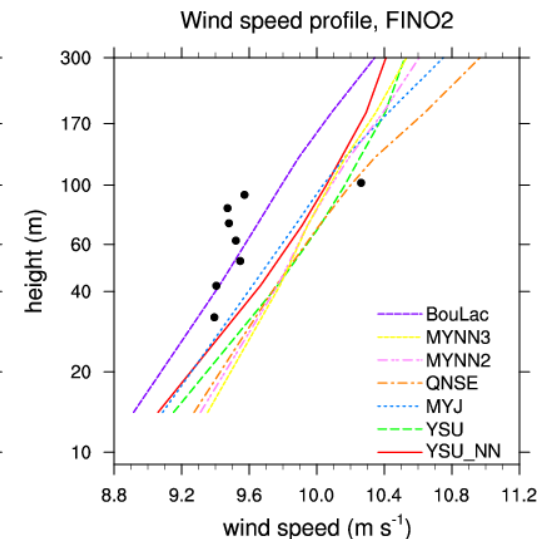
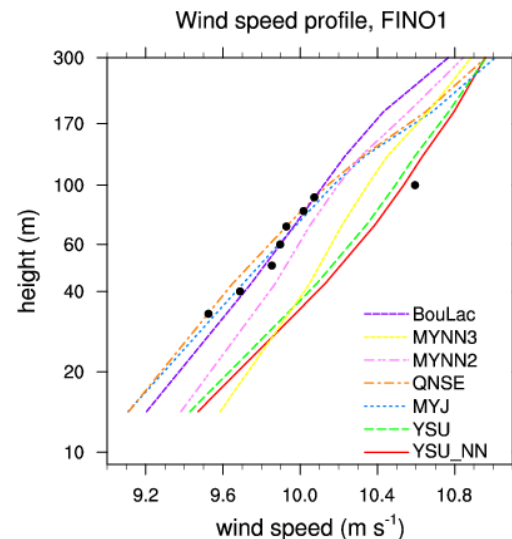
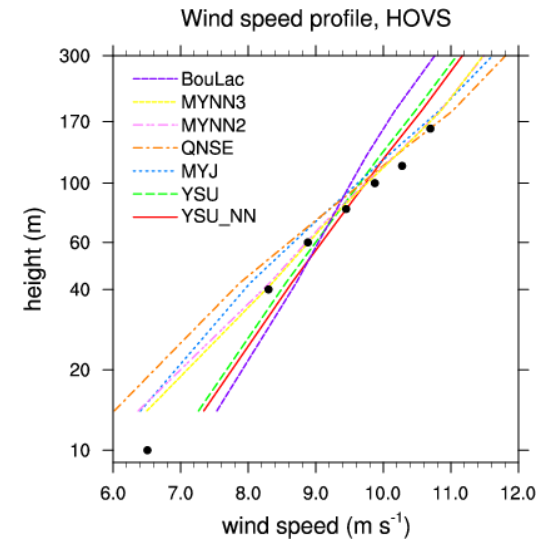
Details of the mesoscale model climatology are important to the coupling strategy



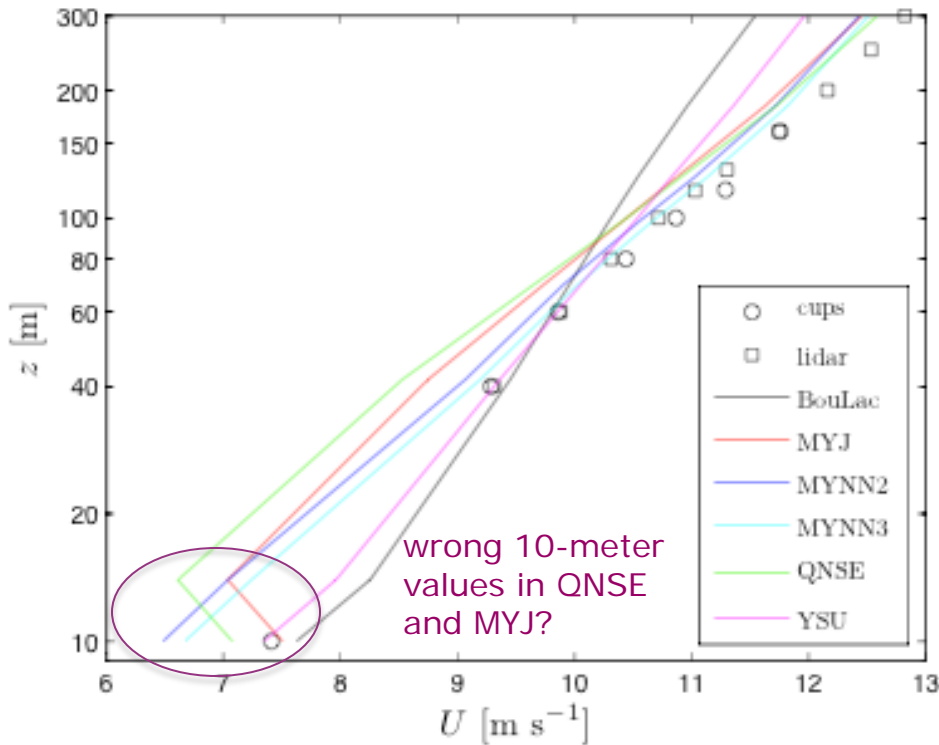
15, 5km dynamical downscaling
(WRF) – CFSR reanalysis

October 2009

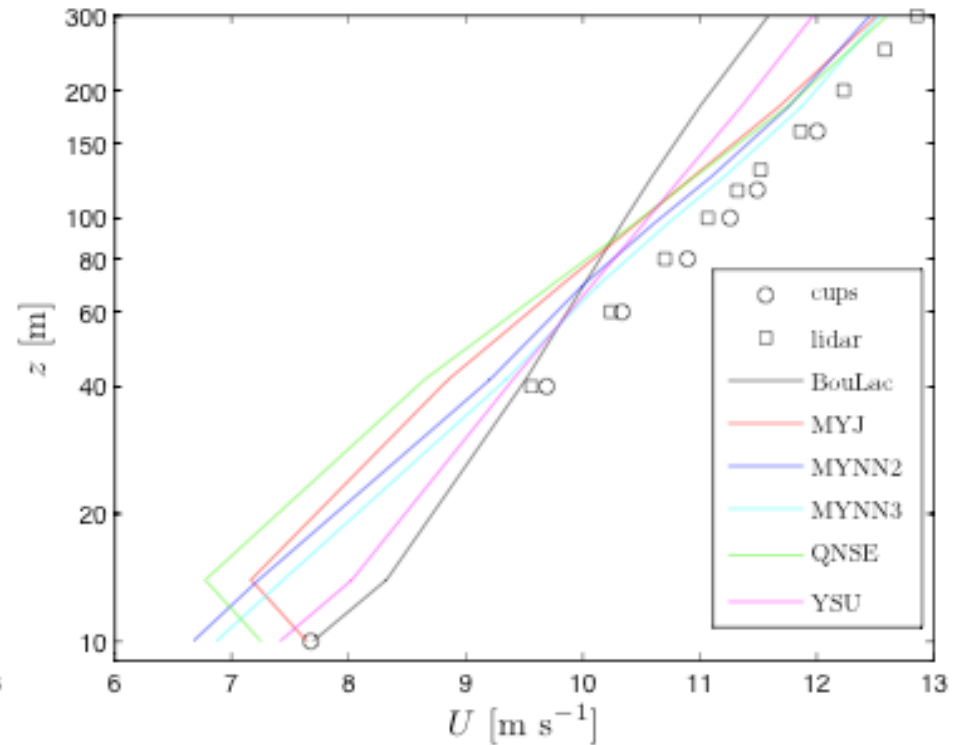
6 boundary layer schemes
(MYNN2, MYNN3, MYJ, QNSE –
KTE schemes, BouLac, YSU –
Non-local schemes)



Further profile verification – Comparison with Cups and Lidar data (Høvsøre, October 2009)

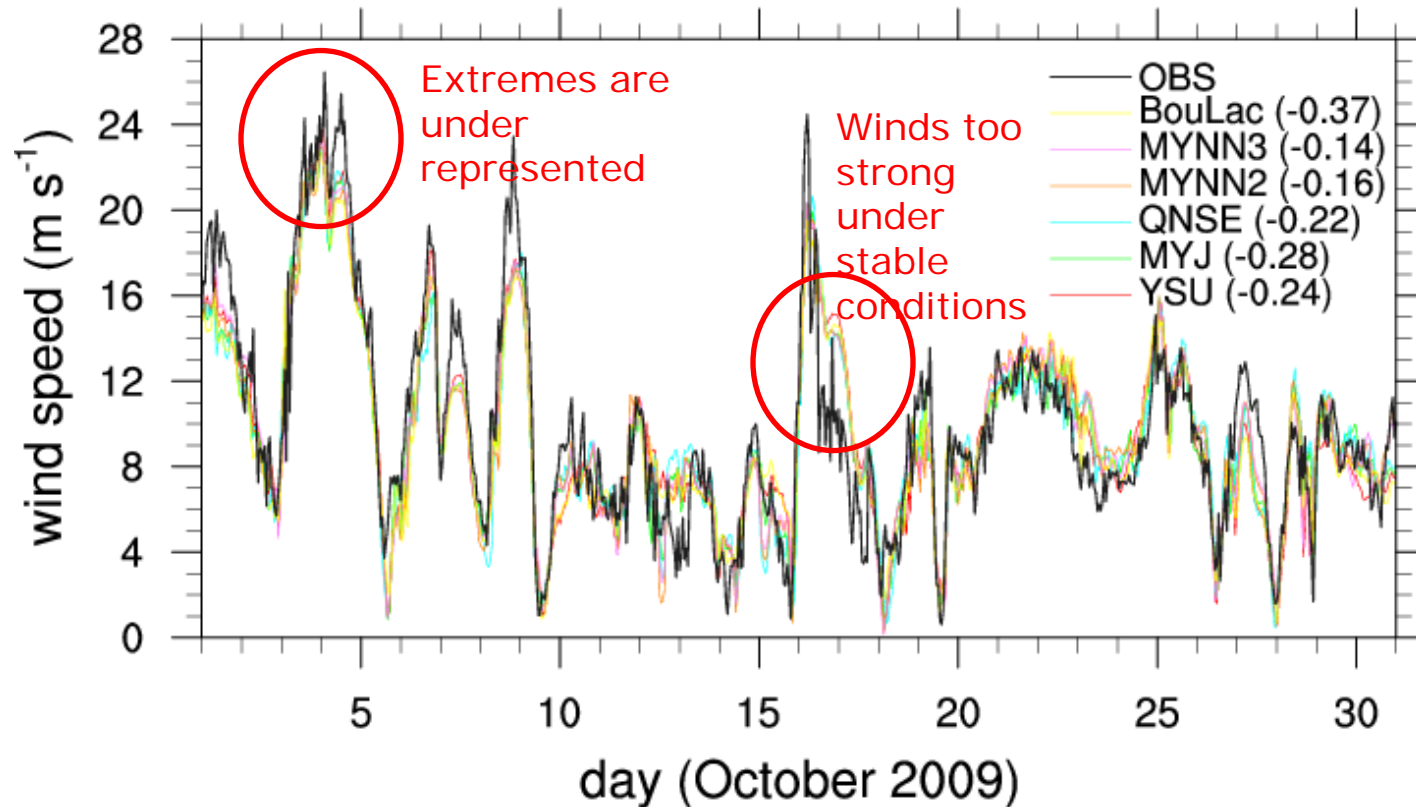


WRF versus wind speed measurements – all sectors



WRF versus wind speed measurements – non-wake sectors

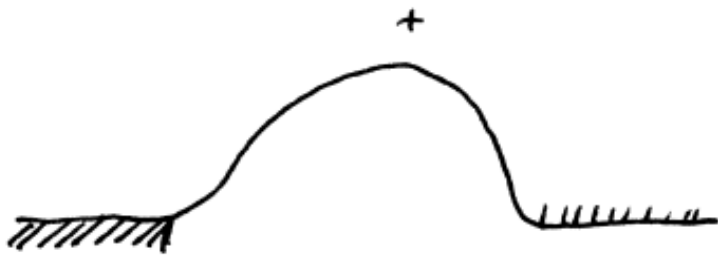
Wind speed, HOVS; height: 100 m



How do we use the knowledge about the errors in the simulation to devise a better coupling strategy?

Mesoscale to microscale coupling: Need for generalization

mesoscale view

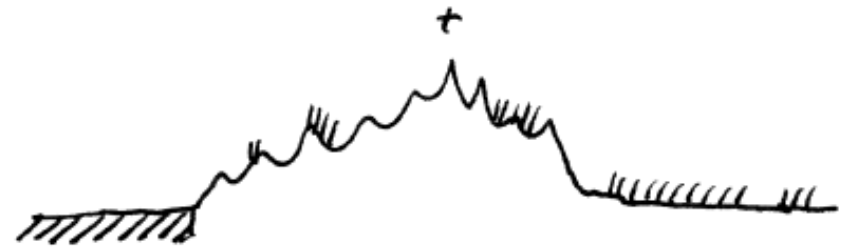


$$h(x) = f(x)$$

$$z_0(x) = r(x)$$

\neq

reality view



$$f(x) \neq g(x)$$

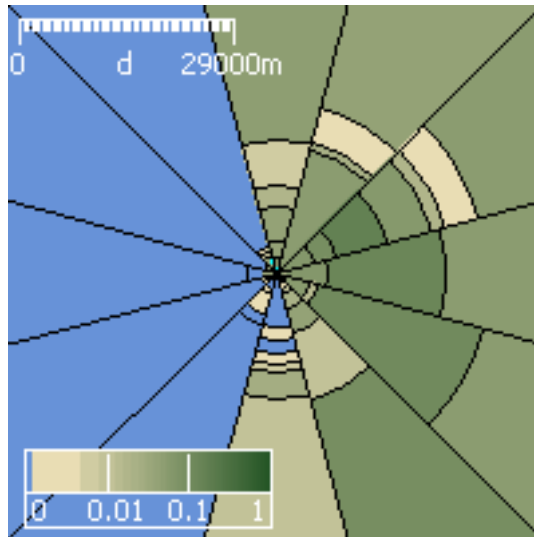
$$r(x) \neq s(x)$$

$$h(x) = g(x)$$

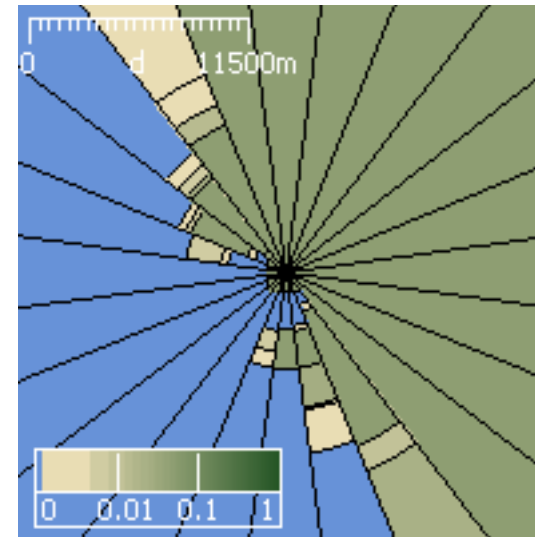
$$z_0(x) = s(x)$$

Høvsøre, Denmark

roughness rose from
high-resolution maps



roughness rose from
WRF land use



To standardize measurements and model values are “corrected” using:

- WAsP speed-up factors (roughness and topography)
- Logarithmic and “geostrophic” wind laws

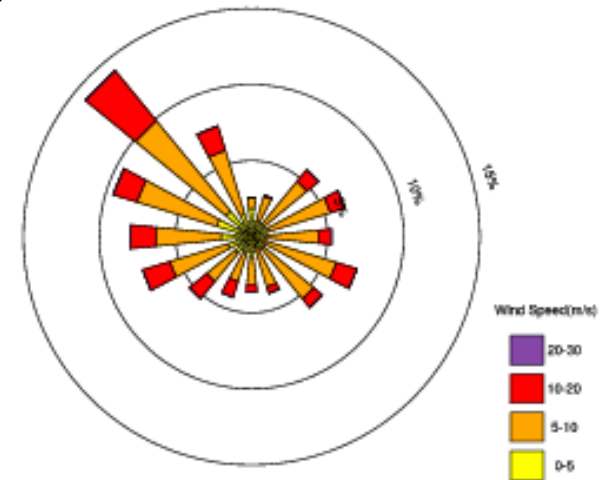
$$u_z = u_{0z} / [(1 + s_0)(1 + s_r)]$$

$$u_* = \frac{\kappa}{u_z} \ln(z/z_0)$$

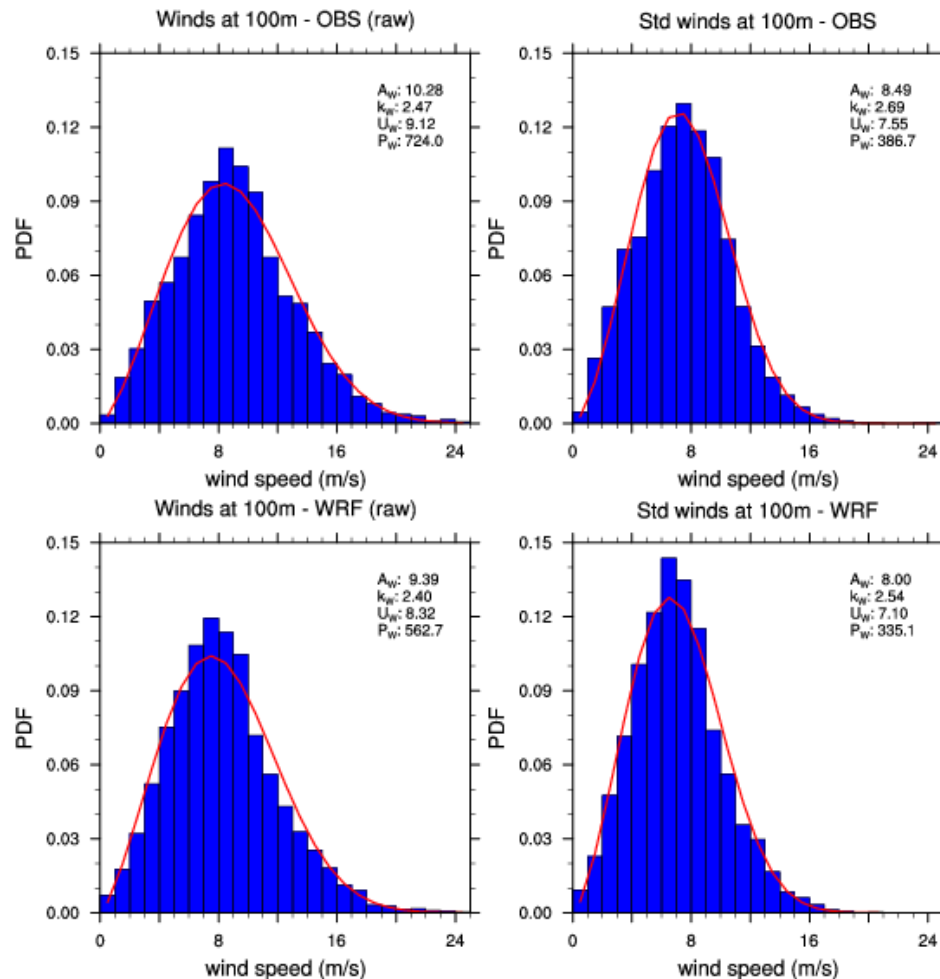
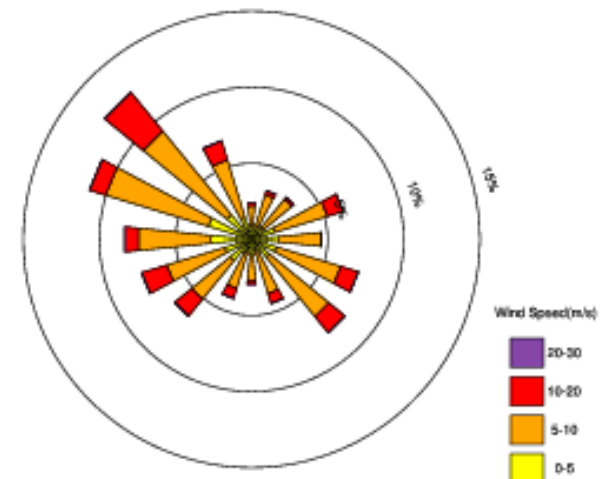
$$G = \frac{u_*}{\kappa} \sqrt{\ln\left(\frac{u_*}{fz_0} - A\right)^2 + B^2}$$

Example of wind generalization for Høvsøre mast measurements and WRF

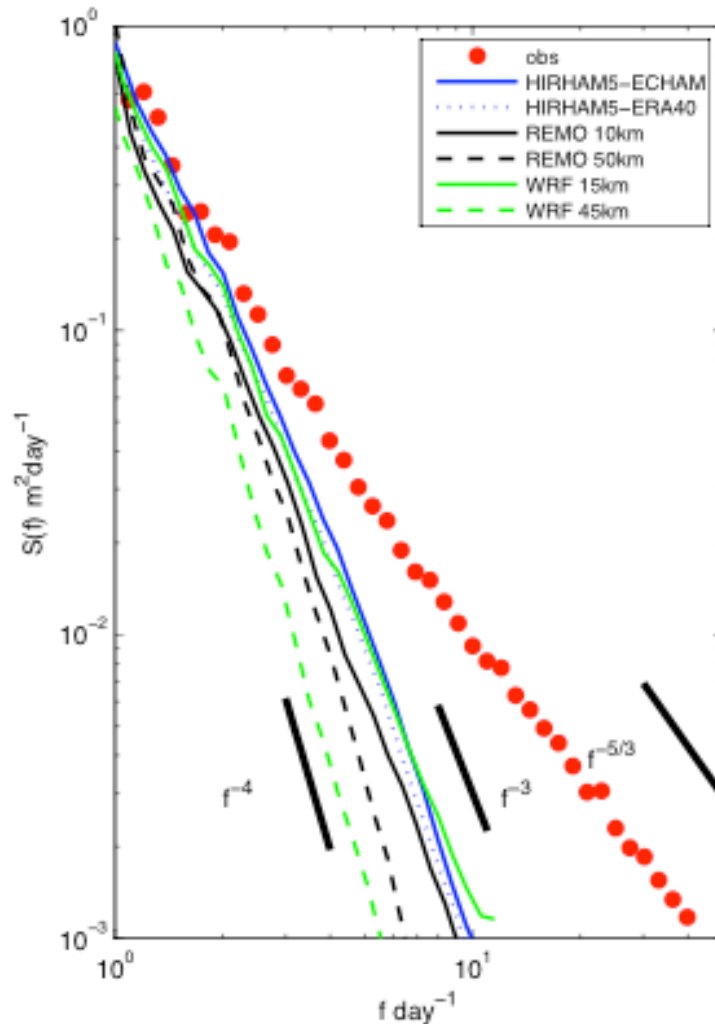
Std winds at 100m - OBS



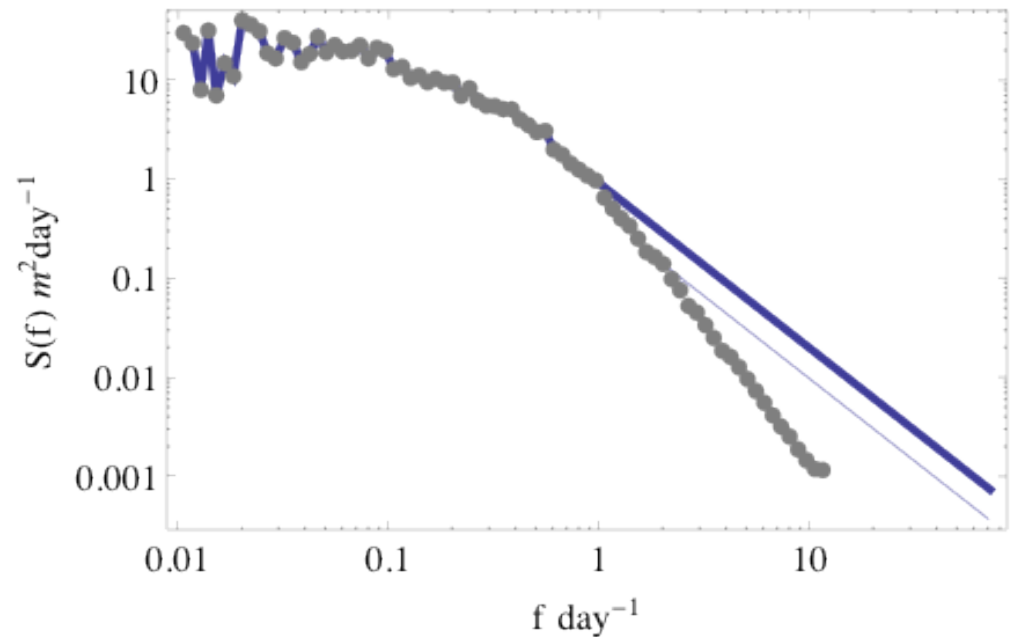
Std winds at 100m - WRF



Other applications – Extreme wind estimation from mesoscale model output



Spectra of wind speed at Horns Rev from observations of various model simulations



Modification of the spectrum of the hourly simulated wind speed

Many remaining issues...

Large-scale to mesoscale coupling:

- nudging (strength, level, fields?) versus re-initialization (how often, spin-period length?)
- length (or sampling strategy) of the simulations – do they capture the interannual (interdecadal) variability?
- what is the adequate spatial resolution – small enough to capture detailed mesoscale structures, large enough for parameterizations to remain valid
- since coupling to microscale – avoid double representing small-scale structures
- ??

Mesoscale to microscale coupling:

- Coupling to linearized models (i.e., WAsP):
 - Generalization works well on wind climatologies – how to expand the concept to include individual observations and model results (need to cover ever more scales...)
 - How do we make use of the deficiencies in the model simulations?
 - ??
- Coupling to more advanced flow models?